



# Planning problems in grids with Distributed Energy Resources

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## **Overview**



- ✓ Introduction
- ✓ Problem Formulation
- ✓ Deterministic approach
- ✓ Evolutionary Algorithms
- ✓ Robust Multi-objective Planning Tool
- √ Test Cases
- ✓ Alternative grids
- ✓ Conclusion





## Typical questions to solve

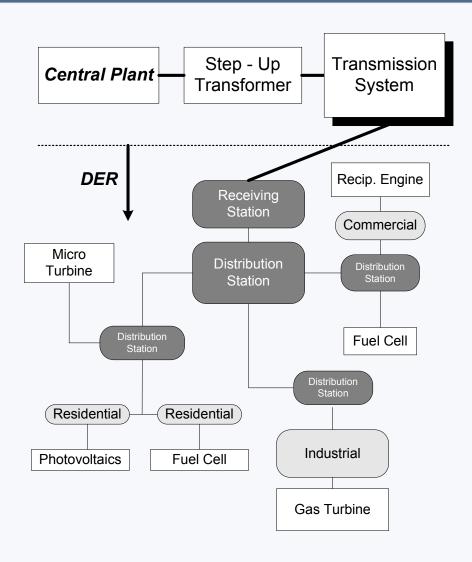


- Customer
  - Which type/size of DG unit will make me the most money?
- DSO
  - How much DG can you put in a grid or feeder, without having to upgrade it?
  - ⇒ Benchmarking?
- Utility
  - When expanding: what is the best next spot for a DG unit?
- Government
  - Where to best spend those subsidies?
  - ⇒ Benchmarking?
- Other
  - ⇒ Is this microgrid self-sustaining?
  - ⇒ economic + technical optimization problems



## Planning problem





## What is the impact of a DER planning scheme?

- Type of DER: renewable, CHP, diesel...
- Size of the unit
- Location in the grid
- Operation mode: peak shaving, market incentives, in coordination with storage...

## Useful for

- Customers: to improve selfsustainability
- Producers: as diversification of production park
- Grid operator or regulator: benchmarking

## Multiple objectives

- Technical: losses, stability, voltage profile, unbalance...
- *Economic:* investment, revenues...
- Risk-averse: deviations due to uncertain future fuel prices, weather conditions, load growth...



## **Problem Formulation**

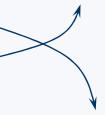


$$V_{\text{node}} \leftarrow \text{flat profile}$$

while  $\varepsilon > \varepsilon_{\text{max}}$ 

$$\text{do} \begin{cases} I_{\text{node}} = f_{\text{load model}} \left( S_{\text{node}}, S_{\text{DER}}, V_{\text{node}} \right) & \text{(a)} \\ I_{\text{line}} = A \cdot I_{\text{node}} & \text{(b)} \\ V_{\text{node}} = V_{\text{grid}} - B \cdot I_{\text{line}} & \text{(c)} \\ \mathcal{E} = g_{\text{error}} \left( S_{\text{node}}, I_{\text{node}}, V_{\text{node}} \right) \end{cases}$$

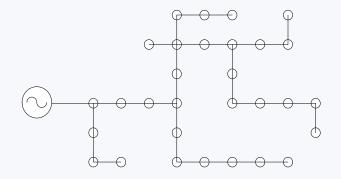
distribution system load flow equations



technical performance

- line losses
- voltage profile
- unbalance

non-linear load model --> iterative procedure 'backward-forward sweeps'



planning problem

- objective: e.g. line loss minimization
- variables: location & size of DER units
- constraints: budget & voltage deviations

nested optimizations in sweep algorithm



## **Planning Issues:** optimization example



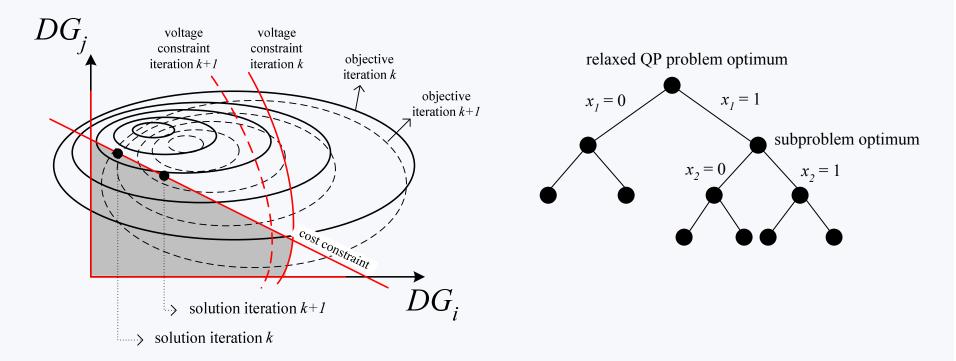
## Main problem (non-linear)



Sequential quadratic optimizations (MIQQ)



Branch and bound search across relaxed QP problems

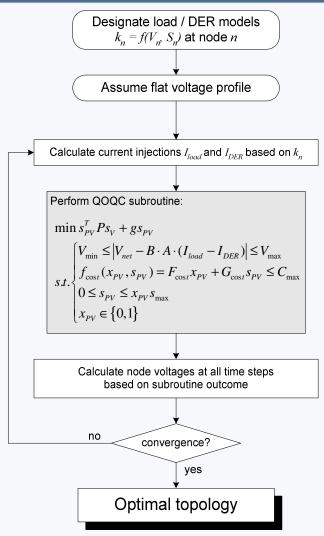




## **Optimization Iteration**



- Quadratic objective
  - RI<sup>2</sup> → multiple nodes, DER types, time frames
- → Quadratic constraint voltage amplitude

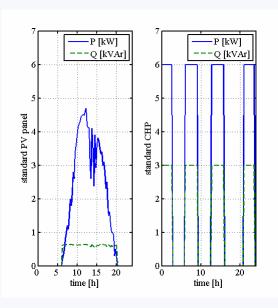


iteration of MIQQ optimizations

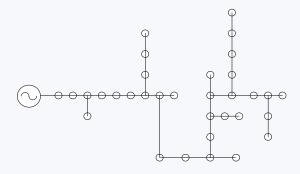


## **Test Case**



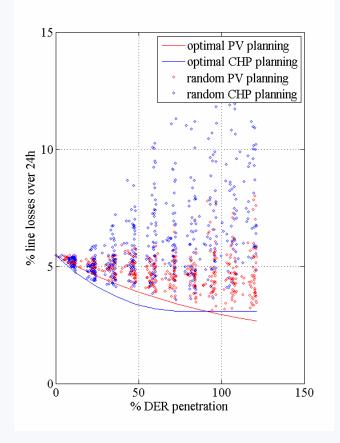


## PV panel and CHP unit placement



IEEE 34-bus radial grid

## optimal versus random placement



## **Conclusions:**

Random deployment: much higher losses low %DER → grid support: best at end of feeders high %DER → node support: relief high loads

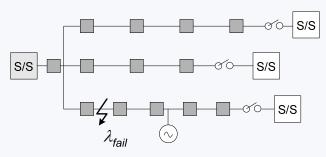


## **Planning Issues**



- Multiple objectives
- Complex objectives/constraints:

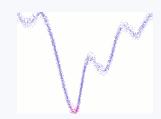
e.g. reliability enhancement, fault currents (urban areas),...



SAIFI, SAIDI, CAIFI, CAIDI, ASAI, ASIFI, ASIDI, ...
(IEEE1366) ⇒ which index? translated to cost?

- What is known?
  - ☑ deterministic

  - ✓ fuzzyVoltage constraints (⇒EN50160?)
  - ✓ uncertain spatial load growth, regulation, ...



→ A robust, multi-objective optimization is needed!

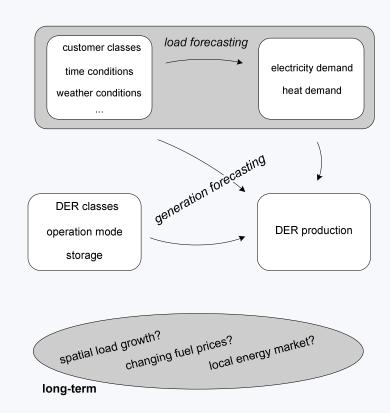


## **Deterministic Optimization**



## Uncertainties remain...

- multiple objectives
  - → acquire trade-off front
- load models?
  - → cause of non-linearity, but not the main uncertainty
- load profiles, weather conditions, load growth, ...?

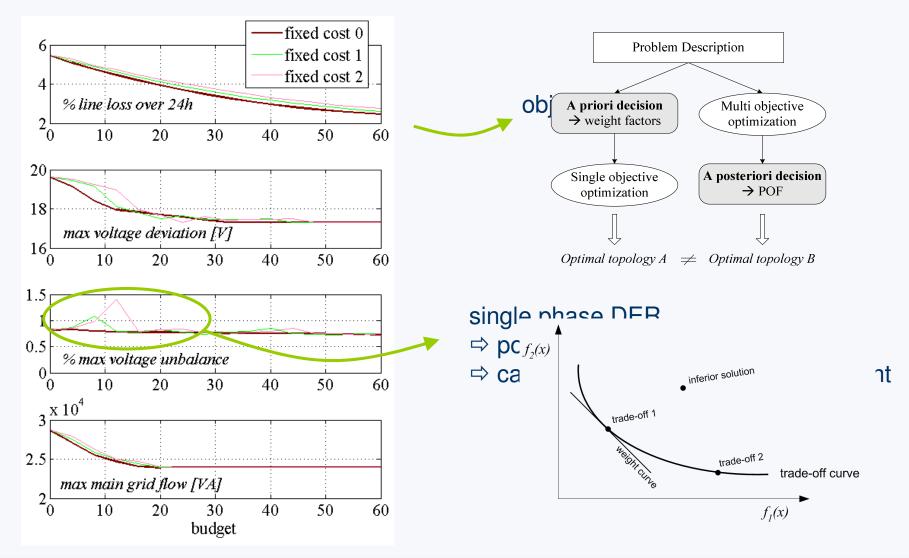


- Deterministic optimization obscures proper planning outcomes
- ⇒ A robust, multi-objective optimization is needed



## **Multiple Objectives**





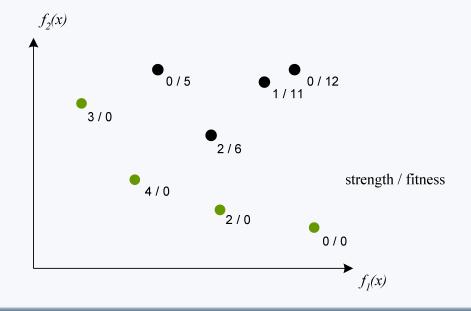


## **Multiple Objectives**



## Pareto Strength

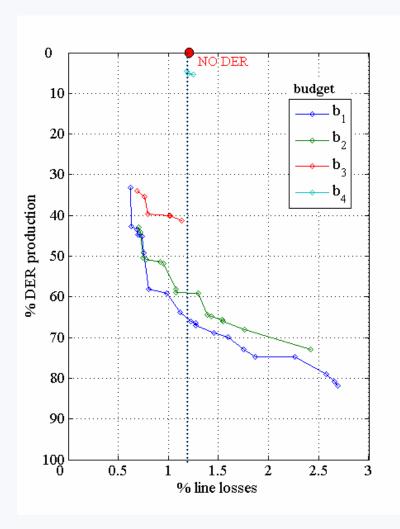
- No use of weight factors
- Search for non-dominated topologies = Pareto Optimal Front
- ⇒ Pareto Strength = #dominated topologies
- ⇒ Fitness = sum of Pareto Strengths of dominating topologies
- (Possible density adjustment)





## **Test cases**





## **Placement**

- PV panels
- micro-CHPs
- wind turbines

## Objectives

- Minimize line losses
- Maximize DER based energy production

## **Constraints**

- ⇒ ∆V
- DER installation budget

$$(b_1 > b_2 > b_3 > b_4)$$





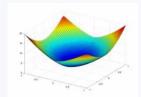
## (Non)-Convex?

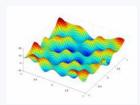
## **Convex**

- local optimum = global optimum
- reliable and efficient algorithms exist
- many apparent non-convex problems can be transformed into convex
  - » line loss minimization
  - » optimal microgrid configuration?

## Non-convex

- application of evolutionary algorithms to search objective space
- non-guaranteed convergence
- high flexibility in function evaluation
  - can handle stochastic data
- multi-objective algorithms
  - acquisition of all trade-offs in one run

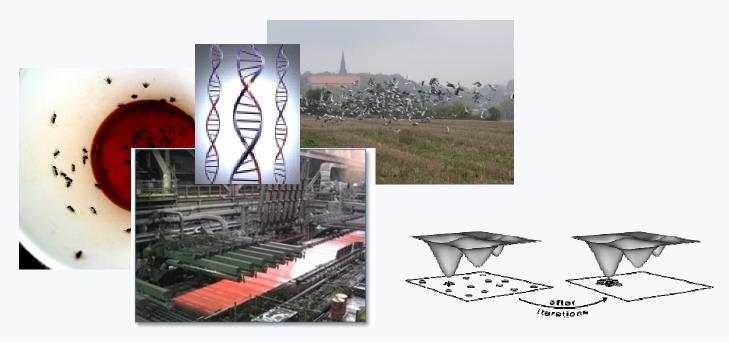








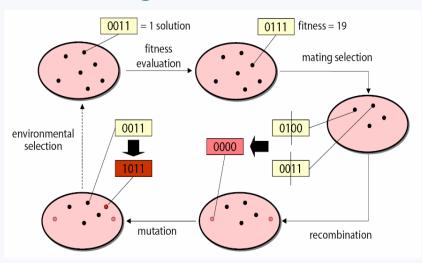








## Genetic Algorithms

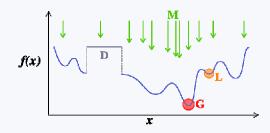


## Simulated Annealing

produce variations j & lower temperature c

$$\mathbb{P}_{c}(accept(j)) = \{ \begin{matrix} f(j) \leq f(i) & 1 \\ f(i) < f(j) & e^{\frac{f(i) - f(j)}{c}} \end{matrix}$$

• Particle Swarm, Ant colony, Tabu search, ...



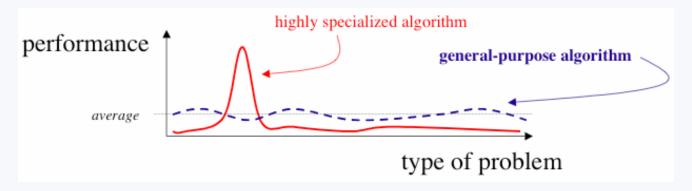
- Population based
- Transitions inspired by laws of biology
- Able to cope with complex problem formulations
  - No derivatives
  - Discontinuous functions
  - Multiple objectives
  - Stochastic formulation
- No guaranteed convergence to global optimum
- Mathematical foundation?







## "There ain't no such thing as a free lunch"



no indication EAs are less reliable planning procedure has to be problem specific !!





## Multiple Objectives

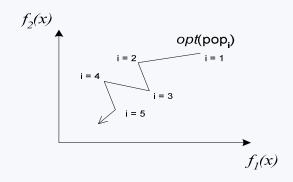
 $\triangleright$  Aggregation  $\varepsilon$  - constrained

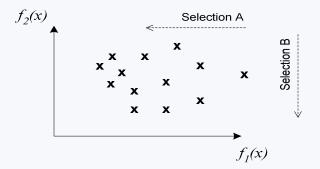
parameter-oriented scaling-dependent

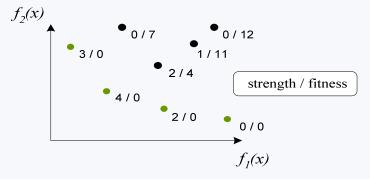
Vector evaluation alternating objectives

**⇒ Pareto optimality**fitness ~ Pareto ranking

set-oriented scaling-independent



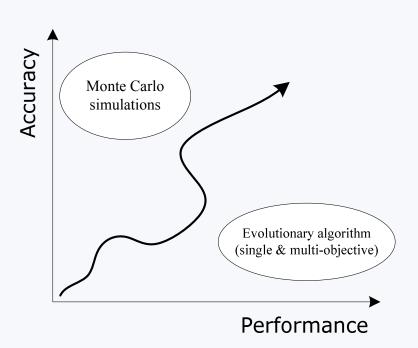


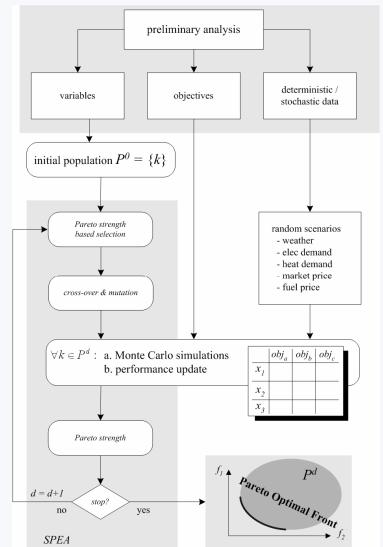




## Stochastic GA



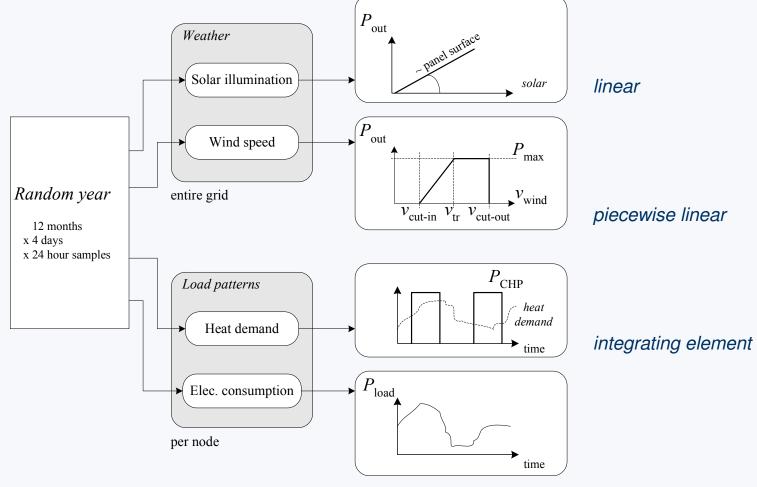






# Stochastic GA input





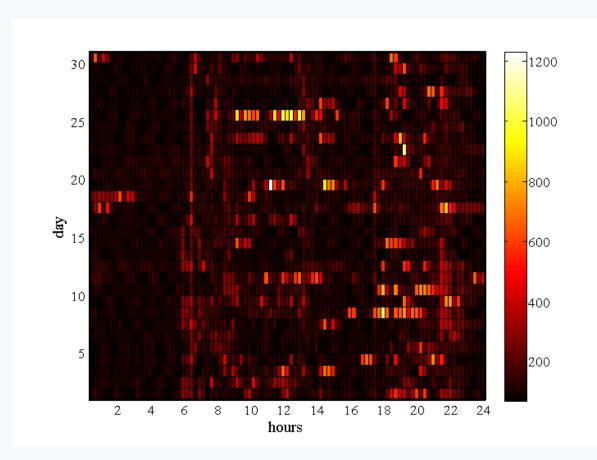
historical data needed!!



# Stochastic GA input



## Modeling of load profiles



# classification year season day hourly hourly hours max/min ...

residential load during May – 15min samples (Wh)



# Stochastic GA input



## Pearson generalized functions

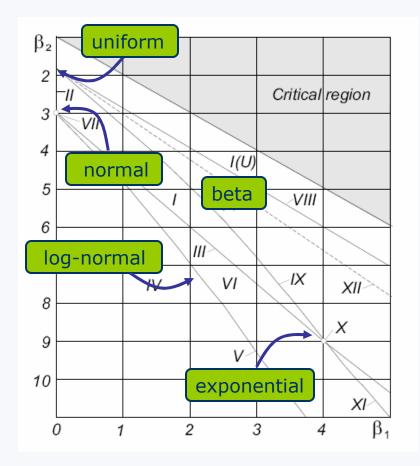
- Probability density function fitting of empirical data
- pdf is characterized by the first 4 moments
- $\circ$  ( $\beta_1$ ,  $\beta_2$ ) indicates most appropriate function

mean value 
$$\mu_1$$
 variance  $\mu_2$  skewness  $\mu_3$  kurtosis  $\mu_4$   $\mu_4$   $\mu_2$   $\mu_4$   $\mu_4$   $\mu_2$   $\mu_4$   $\mu_4$   $\mu_2$   $\mu_4$   $\mu_2$   $\mu_3$   $\mu_4$   $\mu_4$   $\mu_4$   $\mu_2$   $\mu_3$   $\mu_4$   $\mu_$ 

## single residential load

large time-frame: exponential Short time-frame: uniform 3-hour blocks: log-normal

aggregated loads



spatial/temporal load growth?!



# Stochastic GA search pattern



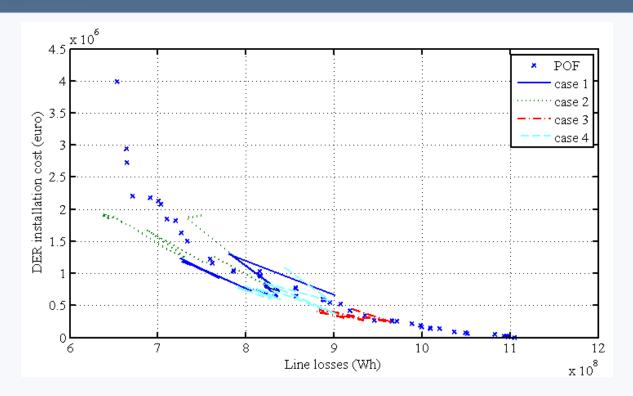
## Heuristics

- Initial population
  - diverse (schema theory)
  - deterministic optimization
- □ Chromosome structure and crossover
  - numbering
  - multiple crossover points
- Mutation
  - not random
  - 3 operators
    - » Creation/deletion
    - » Resizing
    - » Location switching
- Convergence



# Stochastic GA search pattern





## Two objectives

- **⇒ DER cost**
- ⇒ Line losses (NPV) ⇒ DER lifetime? Interest rat

		i = 8%	i = 5%
4	n = 10 years	case 1	case 2
	n = 20 years	case 3	case 4

- Chaotic search of aggregated objective
- Slower convergence of SPEA

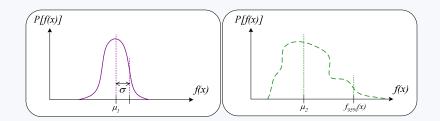


# Stochastic GA search pattern



## **Attributes**

- **□ Deterministic** straightforward
- **□Stochastic** percentile values



- □ Fuzzy
  in final decision making
- □Uncertain → risk scenario approach: e.g. minimal regret



# Stochastic GA output



Decision making strategies

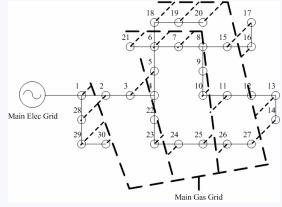
'popular' techniques: e.g. fuzzy logic, game theory (out of scope)

Pareto Front Analysis

## test case:

- » 30 bus electricity grid
- » non-fully deployed gas distribution grid
- » integration of PV panels & micro-CHP units
- » minimize {line losses, electric energy import,
  DER installation cost, gas grid investment}

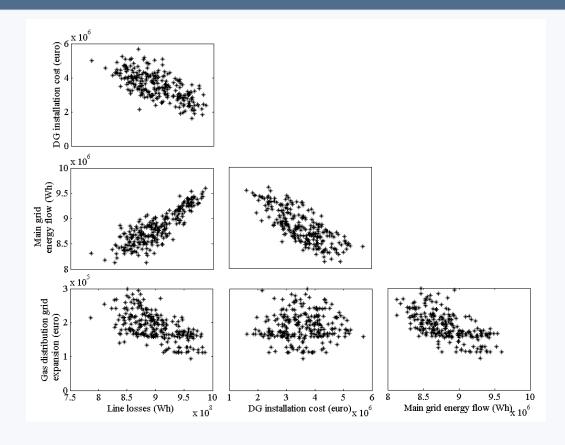
conflicting, related, uncorrelated objectives?





# Stochastic GA output





- two objectives ⇒ simple visualization
- ≥3 ⇒ ?
- Identify strongest trade-offs

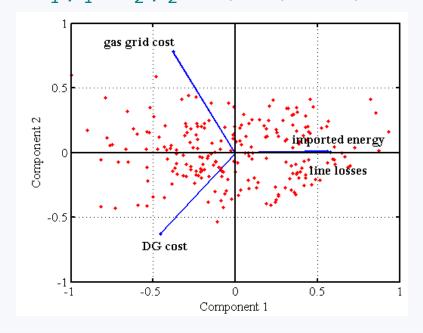


# Stochastic GA output



## Principal Component Analysis

→ Reduce #dimensions of the Pareto set S



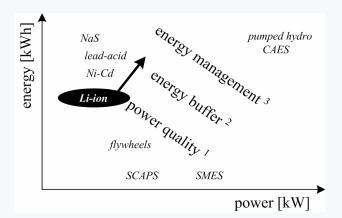
- ✓ 2 out of 4 objectives coincide⇒ 3 real attributes
- ✓ improvement in two attributes= degradation of third

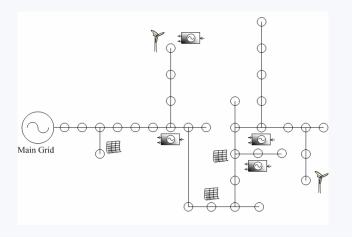


# **Example** storage unit integration



## Integration of small-scale Li-ion battery units in a LV grid with DG





## Variables

battery power & energy rating operation mode: threshold prices at balancing market

## **Objectives**

- ☑ Installation cost (min)
- **☑** Voltage deviation (min)
- **☑** Energy dependence (min)
- ☑ Line losses (min)
- **☑** Conversion losses (min)
- ☑ Revenue at balancing market (max)

## Loads

residential/commercial

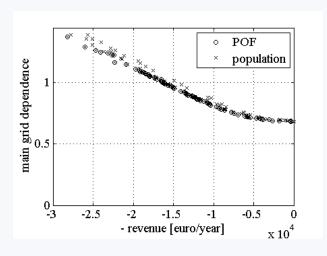
## DG

wind turbines PV panels micro-CHP



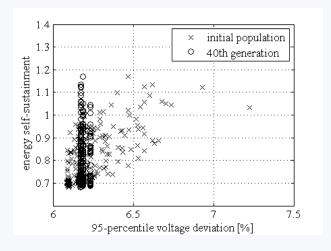
# **Example** storage unit integration







- $\Box$  Energy dependence (= 1  $E_{import}/E_{load}$ )
  - → Wide trade-off
  - → No perfect match DG-load

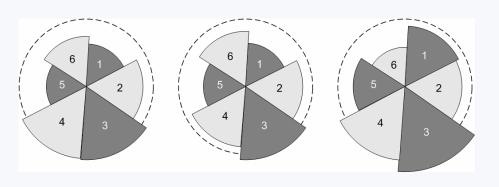


- ☐ Voltage deviations (95-percentile)
- ☐ Energy dependence
  - → correlation
  - → accuracy improvement visible



# **Example** storage unit integration

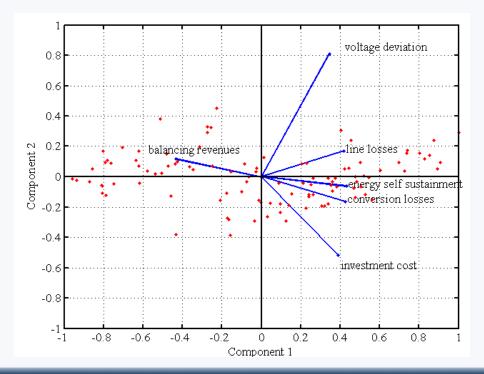




## Petal diagrams

visualize

- multiple
- normalized criteria
- ☐ of specific solutions



## **Principal Component Analysis**



**Storage operation** 

Determine optimal allocation of storage energy for

- □ balancing revenues
- □ ancillary services
- □ energy independence
- **...**



## Conclusion



## Long-term DER planning is analyzed

- Traditional optimization
  - ⇒ deterministic
- A new planning scheme is proposed
  - Multi-objective: search for trade-offs
  - ⇒ Robust: nested MC trials in an evolutionary algorithm
- Convex vs. non-convex: No free lunch





## **Future Work in planning**



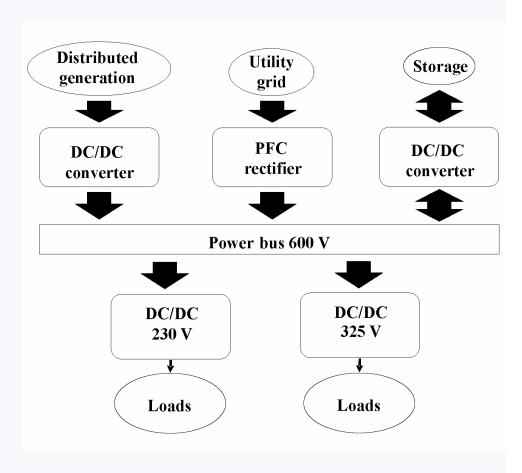
- ⇒ Dispatchable DER sources: link to (short-time) control
  - Gas turbines, Storage units (accepted CIRED paper)
  - Short-term planning vs. long-term planning
- **⇒** Stability issues
  - grid reliability
- Deterministic planning
  - useful for initialization, convergence check



## Extra: are local DC-grids helping?



- Many DG unit have a power electronic frontend performing a DC/AC conversion
- Many loads
   (power supplies)
   contain a AC/DC
- Connect using a DC-connection?





## Why (not) DC?



- Advantages of DC
  - ⇒ Better use of conductors (no skin effect/proximity effect)
  - ⇒ No reactive power
  - Only resistive voltage drops
  - ⇒ Lower transfer losses
  - No standards yet: voltage choice still open

- Disadvantages of DC
  - → More dangerous in case electrocution?
  - ⇒ Hard to interrupt in switchgear
  - No large rotating generators (do we need them?
  - More corrosive
  - Standardized equipment for AC



## DC distribution new?



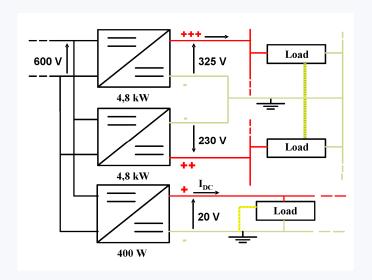
- Edison started that way
- On-board vehicles: DC grid
- Several experiments around the world, e.g. Japan
  - ⇒KUL proposal (IECON06 paper)

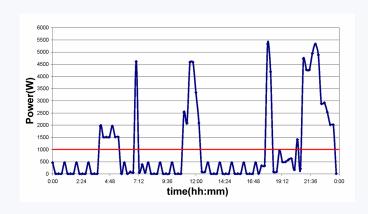


## Assessment for in-house grid



- Model of a house was derived and assessed for several layouts and voltages levels
- Simulations of daily cycles







## **Conclusion on DC-grids**



- Only small increase in efficiency
  - ⇒AC in-house grids already have low losses
- Safety issues can be solved
  - Appropriate grounding
  - Switches still necessary with most loads 'always on'?
- Problem is relatively bad partial load efficiency of DC/DC converters
  - ⇒ Push for advanced designs?





## Thank you for your attention!

Questions?

